

# Carseats-knn.R

karanYsingh

2021-02-15

```
library(ISLR)
```

```
## Warning: package 'ISLR' was built under R version 4.0.3
```

```
data(Carseats)
summary(Carseats)
```

```
##      Sales      CompPrice      Income      Advertising
##  Min.   : 0.000   Min.   : 77   Min.   : 21.00   Min.   : 0.000
## 1st Qu.: 5.390   1st Qu.:115   1st Qu.: 42.75   1st Qu.: 0.000
## Median : 7.490   Median :125   Median : 69.00   Median : 5.000
## Mean   : 7.496   Mean   :125   Mean   : 68.66   Mean   : 6.635
## 3rd Qu.: 9.320   3rd Qu.:135   3rd Qu.: 91.00   3rd Qu.:12.000
## Max.   :16.270   Max.   :175   Max.   :120.00   Max.   :29.000
##      Population      Price      ShelfLoc      Age      Education
##  Min.   : 10.0   Min.   : 24.0   Bad   : 96   Min.   :25.00   Min.   :10.0
## 1st Qu.:139.0   1st Qu.:100.0   Good  : 85   1st Qu.:39.75   1st Qu.:12.0
## Median :272.0   Median :117.0   Medium:219   Median :54.50   Median :14.0
## Mean   :264.8   Mean   :115.8               Mean   :53.32   Mean   :13.9
## 3rd Qu.:398.5   3rd Qu.:131.0               3rd Qu.:66.00   3rd Qu.:16.0
## Max.   :509.0   Max.   :191.0               Max.   :80.00   Max.   :18.0
##      Urban      US
##  No :118   No :142
## Yes:282   Yes:258
##
##
##
##
```

```
names(Carseats)
```

```
## [1] "Sales"      "CompPrice"  "Income"     "Advertising" "Population"
## [6] "Price"      "ShelveLoc"  "Age"        "Education"   "Urban"
## [11] "US"
```

```
library(e1071)
```

```
## Warning: package 'e1071' was built under R version 4.0.3
```

```
library(caTools)
```

```
## Warning: package 'caTools' was built under R version 4.0.3
```

```
library(class)
```

```
## Warning: package 'class' was built under R version 4.0.3
```

```
fix
```

```
## function (x, ...)
## {
##     subx <- substitute(x)
##     if (is.name(subx))
##         subx <- deparse(subx)
##     if (!is.character(subx) || length(subx) != 1L)
##         stop("'fix' requires a name")
##     parent <- parent.frame()
##     if (exists(subx, envir = parent, inherits = TRUE))
##         x <- edit(get(subx, envir = parent), title = subx, ...)
##     else {
##         x <- edit(function() {
##             }, title = subx, ...)
##         environment(x) <- .GlobalEnv
##     }
##     assign(subx, x, envir = .GlobalEnv)
## }
## <bytecode: 0x00000000139fdff0>
## <environment: namespace:utils>
```

```
set.seed(9850)
```

```
gp<-runif(nrow(Carseats))
```

```
gp
```

```
##      [1] 0.7495758825 0.9970860172 0.6520019539 0.4329282779 0.3323124126
##      [6] 0.8654064713 0.1793312423 0.4784936542 0.2957953140 0.6644065792
##     [11] 0.7117703257 0.7801193437 0.1356792178 0.0819188494 0.6316598309
##     [16] 0.5296842461 0.0170833713 0.9474428124 0.2882613896 0.0102437572
##     [21] 0.8897227643 0.0998385579 0.9765785441 0.6263110095 0.8189663987
##     [26] 0.8142171314 0.2289382669 0.3933451127 0.9630645688 0.1585229489
##     [31] 0.3653627359 0.8156728500 0.1604756210 0.3250754566 0.9560772402
##     [36] 0.2218299413 0.2403588612 0.6548423267 0.8333722104 0.2233174085
##     [41] 0.5673194607 0.6456432254 0.0465932356 0.1668617709 0.0956007408
##     [46] 0.2807086881 0.2743727185 0.3060213479 0.4662361774 0.7148338449
##     [51] 0.8190107290 0.4139657072 0.0334808941 0.1631714739 0.6145221728
##     [56] 0.6255910993 0.9879692337 0.5909557175 0.2919432321 0.8480136376
##     [61] 0.2396267867 0.2271126616 0.0140372261 0.2352089230 0.3484869979
##     [66] 0.7520098684 0.3978046870 0.1736333375 0.1154108711 0.0966818000
##     [71] 0.7729893238 0.5798524946 0.5923610253 0.3171202319 0.2654884034
##     [76] 0.7360402378 0.7254672160 0.8663905747 0.7741213529 0.0413182371
##     [81] 0.7172160284 0.6180823920 0.0251395968 0.9586153277 0.2370699421
```

```

## [86] 0.1578509451 0.9192689520 0.3734502334 0.6125246426 0.5803679214
## [91] 0.0404882729 0.6060245500 0.2533204837 0.6174646146 0.6709800093
## [96] 0.4986784160 0.5396321670 0.6206036371 0.7436010945 0.7073883556
## [101] 0.8810440945 0.1403651948 0.0095803072 0.6410322264 0.9191053398
## [106] 0.9689641173 0.1570933873 0.4948440322 0.4536333787 0.2113605556
## [111] 0.8035114724 0.5228704978 0.5376669175 0.9227446702 0.9753406069
## [116] 0.2237336703 0.3711161797 0.0353129525 0.3222257327 0.1571450704
## [121] 0.1685013452 0.0638064316 0.4893020147 0.9682556493 0.4745871460
## [126] 0.2694554001 0.4516987344 0.2202123308 0.9867946801 0.7751187934
## [131] 0.6587832973 0.0880496269 0.8152278706 0.7439878148 0.0836194784
## [136] 0.4659152231 0.7591692442 0.8864776315 0.6038413683 0.1559114396
## [141] 0.7475329281 0.6899246653 0.9861639573 0.5126079635 0.3113563182
## [146] 0.6440915293 0.8044008203 0.1331510283 0.7577468259 0.4200751849
## [151] 0.9882166183 0.5940190239 0.1337920092 0.1093714633 0.4527440837
## [156] 0.4870031353 0.4772156312 0.6883584394 0.1238144503 0.1765312739
## [161] 0.0513233980 0.6495025042 0.6077275414 0.5147005529 0.2526203436
## [166] 0.0743994124 0.0132462864 0.8147595294 0.7346024157 0.9882323125
## [171] 0.7914368890 0.0421418818 0.4482217485 0.9431283800 0.3041717457
## [176] 0.3037171357 0.2458586334 0.1390944780 0.9930689118 0.0179640960
## [181] 0.6333474671 0.3473129433 0.3848697122 0.0925622813 0.3150161311
## [186] 0.3367951058 0.0606662470 0.8249456959 0.8221009444 0.2811661591
## [191] 0.9516206449 0.2198111697 0.3371348912 0.4664305942 0.0453822589
## [196] 0.5599722816 0.3349301305 0.2032854704 0.8286181735 0.6925395546
## [201] 0.9271206164 0.8221291730 0.9972034085 0.6973466040 0.5169499926
## [206] 0.5990863321 0.3223969345 0.4964734735 0.4476255970 0.5216260178
## [211] 0.7556086318 0.3205462527 0.9506412528 0.2985267481 0.0569786283
## [216] 0.0003940333 0.9927308816 0.7886940781 0.5844371154 0.2674444148
## [221] 0.3897148769 0.3091349881 0.0609547086 0.7870363081 0.0268985992
## [226] 0.3048943724 0.1436142863 0.6270551153 0.9310787276 0.1233180955
## [231] 0.9694899826 0.9042756297 0.4780196615 0.9997890533 0.1102453312
## [236] 0.8979925164 0.6166725135 0.8766021957 0.4819052627 0.6744408454
## [241] 0.6150198269 0.5775543137 0.7899192325 0.3725320750 0.1857498381
## [246] 0.2536579652 0.7994820816 0.0724229221 0.4911580693 0.8653213726
## [251] 0.7921922009 0.7603564074 0.7658275897 0.2609684896 0.4687628602
## [256] 0.2254378458 0.8437963750 0.5267864007 0.6994121750 0.2194651333
## [261] 0.1764618107 0.3986164399 0.2646194524 0.0251766567 0.0470231422
## [266] 0.1683996238 0.4098842982 0.0107868451 0.4487984118 0.5636629767
## [271] 0.1780302855 0.0754295127 0.6989828863 0.0783133446 0.5562564055
## [276] 0.4407939913 0.5425982976 0.2262612802 0.4214579179 0.6611826962
## [281] 0.7319047865 0.4298188754 0.8704783071 0.3572077535 0.3513338610
## [286] 0.6648822455 0.7005042701 0.8147210029 0.3996182922 0.4515349315
## [291] 0.2052623765 0.2991150464 0.7789198267 0.8251774590 0.2134868747
## [296] 0.5609124941 0.9681383334 0.8594777016 0.9657609260 0.0412506310
## [301] 0.7548311418 0.4755594810 0.0729731962 0.0331127706 0.5713228611
## [306] 0.2930599167 0.8516822257 0.1701295599 0.3760239882 0.6877743497
## [311] 0.9270636514 0.5851917360 0.8916449999 0.9579640331 0.9715138061
## [316] 0.0018148751 0.9867867955 0.2682508330 0.1255420763 0.6653894021
## [321] 0.6639811567 0.0419312005 0.6743365913 0.2804404830 0.3287259294
## [326] 0.2378482595 0.0931047192 0.9018054479 0.1877670337 0.0448879760
## [331] 0.4912957121 0.7039545635 0.2069241672 0.1909614610 0.5143850462
## [336] 0.0790889347 0.8782816182 0.9176112625 0.7554454068 0.5667540238
## [341] 0.0503111489 0.5667534743 0.6279647511 0.3004350450 0.6171553163
## [346] 0.1975540451 0.6387072350 0.9078744042 0.3736896049 0.1981965557
## [351] 0.8907989997 0.1360701858 0.8814244869 0.2171411105 0.3125700054

```

```
## [356] 0.4010509730 0.1826840122 0.5822353936 0.6359907584 0.0989283440
## [361] 0.6765615689 0.1485178072 0.3771007759 0.5530016017 0.3617152253
## [366] 0.1802198892 0.6171681150 0.8206919797 0.6106209494 0.6188949822
## [371] 0.1965838522 0.7399326509 0.4477081764 0.8405267783 0.7720598341
## [376] 0.4463721714 0.2442423559 0.4129812869 0.9434902302 0.4571786132
## [381] 0.3632806947 0.8853263238 0.9804098252 0.2628147874 0.4995117523
## [386] 0.0783771609 0.9786627586 0.7474930072 0.9742306971 0.6519231112
## [391] 0.5787591410 0.8504999247 0.1858147094 0.5311700099 0.3757693039
## [396] 0.0710409759 0.1168735998 0.2629899543 0.0945292274 0.7763616967
```

```
Carseats<-Carseats[order(gp),]
Carseats
```

##	Sales	CompPrice	Income	Advertising	Population	Price	ShelveLoc	Age	Education
## 216	2.34	116	83	15	170	144	Bad	71	11
## 316	6.39	131	21	8	220	171	Good	29	14
## 103	5.30	113	22	0	57	97	Medium	65	16
## 20	8.73	129	76	16	58	121	Medium	69	12
## 268	5.83	134	82	7	473	112	Bad	51	12
## 167	6.71	119	67	17	151	137	Medium	55	11
## 63	1.82	139	45	0	146	133	Bad	77	17
## 17	7.58	118	32	0	284	110	Good	63	13
## 180	7.78	144	25	3	70	116	Medium	77	18
## 83	11.62	151	83	4	325	139	Good	28	17
## 264	7.77	116	26	6	434	115	Medium	25	17
## 225	4.10	134	82	0	464	141	Medium	48	13
## 304	10.01	133	52	16	290	99	Medium	43	11
## 53	7.91	153	40	3	112	129	Bad	39	18
## 118	8.80	145	53	0	507	119	Medium	41	12
## 91	5.33	115	22	0	491	103	Medium	64	11
## 300	9.40	135	40	17	497	96	Medium	54	17
## 80	9.14	134	67	0	286	90	Bad	41	13
## 322	7.52	123	39	5	499	98	Medium	34	15
## 172	12.49	93	106	12	416	55	Medium	75	15
## 330	11.27	100	54	9	433	89	Good	45	12
## 195	7.23	112	98	18	481	128	Medium	45	11
## 43	10.43	77	69	0	25	24	Medium	50	18
## 265	6.95	128	29	5	324	159	Good	31	15
## 341	7.50	140	29	0	105	91	Bad	43	16
## 161	4.67	111	28	0	486	111	Medium	29	12
## 215	4.83	115	115	3	48	107	Medium	73	18
## 187	8.68	120	51	0	93	86	Medium	46	17
## 223	7.49	136	119	6	178	145	Medium	35	13
## 122	11.67	125	89	10	380	87	Bad	28	10
## 396	12.57	138	108	17	203	128	Good	33	14
## 248	5.04	123	114	0	298	151	Bad	34	16
## 303	5.28	108	77	13	388	110	Bad	74	14
## 166	0.37	147	58	7	100	191	Bad	27	15
## 272	4.55	111	56	0	504	110	Medium	62	16
## 274	10.04	116	106	8	244	86	Medium	58	12
## 386	5.87	131	73	13	455	132	Medium	62	17
## 336	6.18	120	70	15	464	110	Medium	72	15
## 14	10.96	115	28	11	29	86	Good	53	18
## 135	3.67	132	31	0	327	131	Medium	76	16

## 132	6.50	108	69	3	208	94	Medium	77	16
## 184	5.32	118	74	6	426	102	Medium	80	18
## 327	4.69	133	30	0	152	122	Medium	53	17
## 399	5.94	100	79	7	284	95	Bad	50	12
## 45	4.16	85	79	6	325	95	Medium	69	13
## 70	7.99	127	59	0	339	99	Medium	65	12
## 360	3.13	130	62	11	396	130	Bad	66	14
## 22	12.13	134	29	12	239	109	Good	62	18
## 154	5.93	150	36	7	488	150	Medium	25	17
## 235	9.43	115	62	11	289	129	Good	56	16
## 69	13.39	149	69	20	366	134	Good	60	13
## 397	6.14	139	23	3	37	120	Medium	55	11
## 230	11.19	98	104	0	404	72	Medium	27	18
## 159	12.53	142	90	1	189	112	Good	39	10
## 319	10.08	116	72	10	456	130	Good	41	14
## 148	10.51	140	54	9	402	119	Good	41	16
## 153	7.64	128	78	0	341	128	Good	45	13
## 13	3.98	122	35	2	393	136	Medium	62	18
## 352	10.44	124	115	16	458	105	Medium	62	16
## 178	10.48	138	72	0	148	94	Medium	27	17
## 102	6.20	128	93	0	89	118	Medium	34	18
## 227	7.80	119	33	0	245	122	Good	56	14
## 362	8.68	131	25	10	183	104	Medium	56	15
## 140	12.30	146	62	10	310	94	Medium	30	13
## 107	0.16	102	33	0	217	139	Medium	70	18
## 120	7.37	130	94	8	137	128	Medium	64	12
## 86	8.47	125	103	0	304	112	Medium	49	13
## 30	7.81	104	99	15	226	102	Bad	58	17
## 33	6.20	107	32	12	236	137	Good	64	10
## 54	6.92	109	64	13	39	119	Medium	61	17
## 44	4.12	123	42	11	16	134	Medium	59	13
## 266	5.31	130	35	10	402	129	Bad	39	17
## 121	6.87	128	105	11	249	131	Medium	63	13
## 308	5.90	138	92	0	13	120	Bad	61	12
## 68	9.01	126	61	14	152	115	Medium	47	16
## 261	7.67	129	117	8	400	101	Bad	36	10
## 160	9.32	119	60	0	372	70	Bad	30	18
## 271	11.99	119	26	0	284	89	Good	26	10
## 7	6.63	115	105	0	45	108	Medium	71	15
## 366	6.53	154	30	0	122	162	Medium	57	17
## 357	3.58	142	109	0	111	164	Good	72	12
## 245	8.78	130	30	0	391	100	Medium	26	18
## 393	4.53	129	42	13	315	130	Bad	34	13
## 329	3.15	117	66	1	65	111	Bad	55	11
## 334	5.87	136	60	7	303	147	Medium	41	10
## 371	7.68	126	41	22	403	119	Bad	42	12
## 346	4.81	121	68	0	279	149	Good	79	12
## 350	9.32	134	27	18	467	96	Medium	49	14
## 198	2.52	124	61	0	333	138	Medium	76	16
## 291	9.49	107	111	14	400	103	Medium	41	11
## 333	5.74	106	33	20	354	104	Medium	61	12
## 110	8.98	115	65	0	217	90	Medium	60	17
## 295	12.66	148	76	3	126	99	Good	60	11
## 354	9.45	107	67	12	430	92	Medium	35	12

## 260	5.12	123	36	10	467	100	Bad	74	11
## 192	6.67	156	42	13	170	173	Good	74	14
## 128	6.52	125	48	3	192	116	Medium	51	14
## 36	11.07	131	84	11	29	96	Medium	44	17
## 40	3.24	130	60	0	144	138	Bad	38	10
## 116	8.54	139	35	0	95	129	Medium	42	13
## 256	7.71	123	81	8	198	81	Bad	80	15
## 278	7.80	136	48	12	326	125	Medium	36	16
## 62	7.32	105	32	0	358	107	Medium	26	13
## 27	8.33	107	115	11	496	131	Good	50	11
## 64	8.47	119	88	10	170	101	Medium	61	13
## 85	2.23	111	25	0	52	121	Bad	43	18
## 326	11.70	144	69	11	131	104	Medium	47	11
## 61	8.32	122	102	19	469	123	Bad	29	13
## 37	8.89	122	76	0	270	100	Good	60	18
## 377	16.27	141	60	19	319	92	Good	44	11
## 177	5.61	138	107	9	480	154	Medium	47	11
## 165	8.22	148	64	0	58	141	Medium	27	13
## 93	4.53	114	113	0	97	125	Medium	29	12
## 246	10.00	114	43	0	199	88	Good	57	10
## 254	5.64	124	24	5	288	122	Medium	57	12
## 384	9.35	98	117	0	76	68	Medium	63	10
## 398	7.41	162	26	12	368	159	Medium	40	18
## 263	6.37	120	77	15	86	132	Medium	48	18
## 75	6.20	150	68	5	125	136	Medium	64	13
## 220	10.62	116	79	19	359	116	Good	58	17
## 318	6.41	142	30	0	472	136	Good	80	15
## 126	9.34	89	78	0	181	49	Medium	43	15
## 47	12.44	127	90	14	16	70	Medium	48	15
## 324	10.36	107	105	18	428	103	Medium	34	12
## 46	4.56	141	63	0	168	135	Bad	44	12
## 190	12.11	118	117	18	509	104	Medium	26	15
## 19	13.91	110	110	0	408	68	Good	46	17
## 59	5.42	103	93	15	188	103	Bad	74	16
## 306	8.03	115	29	26	394	132	Medium	33	13
## 9	6.54	132	110	0	108	124	Medium	76	10
## 214	8.23	149	84	5	220	139	Medium	33	10
## 292	6.64	118	70	0	106	89	Bad	39	17
## 344	5.99	117	42	10	371	121	Bad	26	14
## 176	7.54	115	89	0	38	122	Medium	25	12
## 175	0.00	139	24	0	358	185	Medium	79	15
## 226	6.68	107	25	0	412	82	Bad	36	14
## 48	4.38	126	98	0	173	108	Bad	55	16
## 222	6.43	124	44	0	125	107	Medium	80	11
## 145	9.09	132	68	0	264	123	Good	34	11
## 355	5.30	133	31	1	80	145	Medium	42	18
## 185	9.95	132	33	7	35	97	Medium	60	11
## 74	12.61	118	90	10	54	104	Good	31	11
## 212	9.39	117	118	14	445	120	Medium	32	15
## 119	7.57	112	88	2	243	99	Medium	62	11
## 207	4.97	162	67	0	27	160	Medium	77	17
## 34	8.77	114	38	13	317	128	Good	50	16
## 325	2.66	136	65	4	133	150	Bad	53	13
## 5	4.15	141	64	3	340	128	Bad	38	13

## 197	4.10	130	28	6	410	133	Bad	72	16
## 186	10.07	130	100	11	449	107	Medium	64	10
## 193	7.56	108	26	0	408	93	Medium	56	14
## 182	7.43	121	83	0	79	91	Medium	68	11
## 65	7.80	100	67	12	184	104	Medium	32	16
## 285	6.97	106	46	11	414	96	Bad	79	17
## 284	5.36	135	110	0	112	117	Medium	80	16
## 365	10.50	122	21	16	488	131	Good	30	14
## 381	9.64	106	64	10	17	89	Medium	68	17
## 31	13.55	125	94	0	447	89	Good	30	12
## 117	5.08	135	75	0	202	128	Medium	80	10
## 244	7.82	124	25	13	87	110	Medium	57	10
## 88	11.70	131	67	7	272	126	Good	54	16
## 349	12.57	132	102	20	459	107	Good	49	11
## 395	5.35	130	58	19	366	139	Bad	33	16
## 309	9.24	126	80	19	436	126	Medium	52	10
## 363	5.25	131	55	0	26	110	Bad	79	12
## 183	4.74	137	60	4	230	140	Bad	25	13
## 221	10.59	131	120	15	262	124	Medium	30	10
## 28	5.27	98	118	0	19	107	Medium	64	17
## 67	8.85	127	92	0	508	91	Medium	56	18
## 262	5.71	121	42	4	188	118	Medium	54	15
## 289	6.98	116	40	0	74	97	Medium	76	15
## 356	7.02	130	100	0	306	146	Good	42	11
## 267	9.10	128	93	12	343	112	Good	73	17
## 378	6.81	132	61	0	263	125	Medium	41	12
## 52	4.42	121	90	0	150	108	Bad	75	16
## 150	11.48	121	120	13	140	87	Medium	56	11
## 279	7.22	114	113	2	129	151	Good	40	15
## 282	11.19	122	69	7	303	105	Good	45	16
## 4	7.40	117	100	4	466	97	Medium	55	14
## 276	6.67	107	119	11	210	132	Medium	53	11
## 376	7.90	132	46	4	206	124	Medium	73	11
## 209	7.78	86	54	0	497	64	Bad	33	12
## 373	7.80	121	50	0	508	98	Medium	65	11
## 173	9.03	104	102	13	123	110	Good	35	16
## 269	6.53	123	57	0	66	105	Medium	39	11
## 290	8.75	143	77	25	448	156	Medium	43	17
## 127	11.27	153	68	2	60	133	Good	59	16
## 155	6.89	129	69	10	289	110	Medium	50	16
## 109	3.47	107	79	2	488	103	Bad	65	16
## 380	5.81	125	111	0	404	107	Bad	54	15
## 136	6.44	96	94	14	384	120	Medium	36	18
## 49	3.91	116	52	0	349	98	Bad	69	18
## 194	13.28	139	70	7	71	96	Good	61	10
## 255	9.58	108	104	23	353	129	Good	37	17
## 125	8.87	131	113	0	181	120	Good	63	14
## 302	7.41	99	93	0	198	87	Medium	57	16
## 157	7.49	146	34	0	220	157	Good	51	16
## 233	13.14	137	80	10	24	105	Good	61	15
## 8	11.85	136	81	15	425	120	Good	67	10
## 239	7.36	121	24	0	200	133	Good	73	13
## 156	7.71	98	72	0	59	69	Medium	65	16
## 123	6.88	119	100	5	45	108	Medium	75	10

## 249	5.36	111	52	0	12	101	Medium	61	11
## 331	4.99	122	59	0	501	112	Bad	32	14
## 108	8.55	134	107	0	104	108	Medium	60	12
## 208	8.19	111	105	0	466	97	Bad	61	10
## 96	5.58	134	25	10	237	148	Medium	59	13
## 385	12.85	123	37	15	348	112	Good	28	12
## 144	0.53	122	88	7	36	159	Bad	28	17
## 335	7.63	93	117	9	489	83	Bad	42	13
## 164	5.68	130	64	0	40	106	Bad	39	17
## 205	8.74	155	80	0	237	124	Medium	37	14
## 210	3.02	98	21	11	326	90	Bad	76	11
## 112	6.62	132	118	12	272	151	Medium	43	14
## 258	8.67	125	62	14	477	112	Medium	80	13
## 16	8.71	149	95	5	400	144	Medium	76	18
## 394	5.57	109	51	10	26	120	Medium	30	17
## 113	6.67	116	99	5	298	125	Good	62	12
## 97	9.48	147	42	10	407	132	Good	73	16
## 277	6.93	135	69	14	296	130	Medium	73	15
## 364	10.26	111	75	1	377	108	Good	25	12
## 275	7.22	135	93	2	67	119	Medium	34	11
## 196	4.19	117	93	4	420	112	Bad	66	11
## 296	4.21	118	35	14	502	137	Medium	79	10
## 270	5.01	159	69	0	438	166	Medium	46	17
## 342	7.38	98	120	0	268	93	Medium	72	10
## 340	11.54	134	44	4	219	126	Good	44	15
## 41	2.07	119	98	0	18	126	Bad	73	17
## 305	11.93	123	98	12	408	134	Good	29	10
## 242	12.01	136	63	0	160	94	Medium	38	12
## 391	5.47	108	75	9	61	111	Medium	67	12
## 72	6.50	148	51	16	148	150	Medium	58	17
## 90	7.95	128	66	3	493	119	Medium	45	16
## 358	13.36	103	73	3	276	72	Medium	34	15
## 219	9.70	138	61	12	156	120	Medium	25	14
## 312	6.15	146	68	12	328	132	Bad	51	14
## 58	0.91	93	91	0	22	117	Bad	75	11
## 73	5.52	115	45	0	432	116	Medium	25	15
## 152	10.77	111	58	17	407	103	Good	75	17
## 206	5.68	113	22	1	317	132	Medium	28	12
## 139	10.27	125	103	12	371	109	Medium	44	10
## 92	4.81	97	46	11	267	107	Medium	80	15
## 163	3.63	122	74	0	424	149	Medium	51	13
## 369	10.71	109	22	10	348	79	Good	74	14
## 89	6.56	117	42	7	144	111	Medium	62	10
## 55	4.90	134	103	13	25	144	Medium	76	17
## 241	10.31	159	80	0	362	121	Medium	26	18
## 237	9.32	141	34	16	361	108	Medium	69	10
## 345	8.43	138	80	0	108	126	Good	70	13
## 367	5.98	124	56	11	447	134	Medium	53	12
## 94	8.86	145	30	0	67	104	Medium	55	17
## 82	7.52	116	72	0	237	128	Good	70	13
## 370	10.26	135	100	22	463	122	Medium	36	14
## 98	7.45	161	82	5	287	129	Bad	33	16
## 56	6.85	143	81	5	60	154	Medium	61	18
## 24	5.87	121	31	0	292	109	Medium	79	10



## 228	8.69	113	64	10	68	101	Medium	57	16
## 343	7.81	137	102	13	422	118	Medium	71	10
## 15	11.17	107	117	11	148	118	Good	52	18
## 181	4.94	137	112	15	434	149	Bad	66	13
## 359	4.17	123	96	10	71	118	Bad	69	11
## 347	8.97	132	107	0	144	125	Medium	33	13
## 104	5.07	123	91	0	334	96	Bad	78	17
## 146	8.77	144	63	11	27	117	Medium	47	17
## 42	7.96	157	53	0	403	124	Bad	58	16
## 162	2.93	143	21	5	81	160	Medium	67	12
## 390	8.44	128	42	8	328	107	Medium	35	12
## 3	10.06	113	35	10	269	80	Medium	59	12
## 38	4.95	121	41	5	412	110	Medium	54	10
## 131	8.41	94	84	13	497	77	Medium	51	12
## 280	3.42	141	57	13	376	158	Medium	64	18
## 321	5.86	136	70	12	171	152	Medium	44	18
## 10	4.69	132	113	0	131	124	Medium	76	17
## 286	7.60	146	26	11	261	131	Medium	39	10
## 320	6.97	127	45	19	459	129	Medium	57	11
## 95	8.39	115	97	5	134	84	Bad	55	11
## 323	9.16	140	50	10	300	139	Good	60	15
## 240	3.89	123	105	0	149	118	Bad	62	16
## 361	8.77	118	86	7	265	114	Good	52	15
## 310	11.18	131	111	13	33	80	Bad	68	18
## 158	10.21	121	58	8	249	90	Medium	48	13
## 142	6.53	140	42	0	331	131	Bad	28	15
## 200	6.42	122	88	5	335	126	Medium	64	14
## 204	2.05	131	82	0	132	157	Bad	25	14
## 273	12.98	113	33	0	14	63	Good	38	12
## 259	3.47	108	38	0	251	81	Bad	72	14
## 287	7.53	117	118	11	429	113	Medium	67	18
## 332	10.10	135	63	15	213	134	Medium	32	10
## 100	4.88	121	47	3	220	107	Bad	56	16
## 11	9.01	121	78	9	150	100	Bad	26	10
## 50	10.61	157	93	0	51	149	Good	32	17
## 81	8.01	113	100	16	353	79	Bad	68	11
## 77	10.64	102	87	10	346	70	Medium	64	15
## 281	2.86	121	86	10	496	145	Bad	51	10
## 169	7.30	129	89	0	425	117	Medium	45	10
## 76	8.55	88	111	23	480	92	Bad	36	16
## 372	9.08	152	81	0	191	126	Medium	54	16
## 99	12.49	122	77	24	382	127	Good	36	16
## 134	7.62	132	98	2	265	97	Bad	62	12
## 388	8.67	142	73	14	238	115	Medium	73	14
## 141	6.03	133	60	10	277	129	Medium	45	18
## 1	9.50	138	73	11	276	120	Bad	42	17
## 66	4.90	122	26	0	197	128	Medium	55	13
## 301	8.57	116	78	1	158	99	Medium	45	11
## 339	5.97	112	24	0	164	101	Medium	45	11
## 211	4.36	125	41	2	357	123	Bad	47	14
## 149	7.56	110	119	0	384	97	Medium	72	14
## 137	5.17	131	75	0	10	120	Bad	31	18
## 252	3.72	139	111	5	310	132	Bad	62	13
## 253	8.31	133	97	0	70	117	Medium	32	16

## 375	9.44	131	47	7	90	118	Medium	47	12
## 71	9.46	89	81	15	237	99	Good	74	12
## 79	4.43	134	48	1	139	145	Medium	65	12
## 130	4.47	143	120	7	279	147	Bad	40	10
## 400	9.71	134	37	0	27	120	Good	49	16
## 293	11.82	113	66	16	322	74	Good	76	15
## 12	11.96	117	94	4	503	94	Good	50	13
## 224	3.45	110	45	9	276	125	Medium	62	14
## 218	4.34	106	44	0	481	111	Medium	70	14
## 243	4.68	124	46	0	199	135	Medium	52	14
## 171	8.01	128	39	12	356	118	Medium	71	10
## 251	9.16	137	105	10	435	156	Good	72	14
## 247	6.90	120	56	20	266	90	Bad	78	18
## 111	9.00	128	62	7	125	116	Medium	43	14
## 147	3.90	114	83	0	412	131	Bad	39	14
## 26	14.90	139	32	0	176	82	Good	54	11
## 288	6.88	95	44	4	208	72	Bad	44	17
## 168	6.71	106	73	0	216	93	Medium	60	13
## 133	9.54	125	87	9	232	136	Good	72	10
## 32	8.25	136	58	16	241	131	Medium	44	18
## 25	10.14	145	119	16	294	113	Bad	42	12
## 51	1.42	99	32	18	341	108	Bad	80	16
## 368	14.37	95	106	0	256	53	Good	52	17
## 189	8.07	116	37	0	426	90	Medium	76	15
## 202	5.94	138	83	0	139	134	Medium	54	18
## 188	6.03	117	32	0	142	96	Bad	62	17
## 294	11.28	123	84	0	74	89	Good	59	10
## 199	3.62	112	80	5	500	128	Medium	69	10
## 39	6.59	109	73	0	454	102	Medium	65	15
## 374	5.58	137	71	0	402	116	Medium	78	17
## 257	4.20	147	40	0	277	144	Medium	73	10
## 60	5.21	118	71	4	148	114	Medium	80	13
## 392	6.10	153	63	0	49	124	Bad	56	16
## 307	4.78	131	32	1	85	133	Medium	48	12
## 298	3.07	118	83	13	276	104	Bad	75	10
## 250	5.05	125	67	0	86	117	Bad	65	11
## 6	10.81	124	113	13	501	72	Bad	78	16
## 78	7.70	118	71	12	44	89	Medium	67	18
## 283	7.74	150	96	0	80	154	Good	61	11
## 238	9.62	151	28	8	499	135	Medium	48	10
## 337	5.17	138	35	6	60	143	Bad	28	18
## 101	4.11	113	69	11	94	106	Medium	76	12
## 353	13.44	133	103	14	288	122	Good	61	17
## 382	3.90	124	65	21	496	151	Bad	77	13
## 138	6.52	128	42	0	436	118	Medium	80	11
## 21	6.41	125	90	2	367	131	Medium	35	18
## 351	8.64	111	101	17	266	91	Medium	63	17
## 313	6.80	137	117	5	337	135	Bad	38	10
## 236	5.53	126	32	8	95	132	Medium	50	17
## 328	6.23	112	38	17	316	104	Medium	80	16
## 232	8.09	132	69	0	123	122	Medium	27	11
## 348	6.88	96	39	0	161	112	Good	27	14
## 338	8.61	130	38	0	283	102	Medium	80	15
## 105	4.62	121	96	0	472	138	Medium	51	12

## 87	8.70	150	84	9	432	134	Medium	64	15
## 114	6.01	131	29	11	335	127	Bad	33	12
## 311	9.53	175	65	29	419	166	Medium	53	12
## 201	5.56	144	92	0	349	146	Medium	62	12
## 229	5.40	149	73	13	381	163	Bad	26	11
## 174	6.38	135	91	5	207	128	Medium	66	18
## 379	6.11	133	88	3	105	119	Medium	79	12
## 18	12.29	147	74	13	251	131	Good	52	10
## 213	12.04	145	69	19	501	105	Medium	45	11
## 191	8.79	130	37	13	297	101	Medium	37	13
## 35	2.67	115	54	0	406	128	Medium	42	17
## 314	9.33	103	81	3	491	54	Medium	66	13
## 84	4.42	109	36	7	468	94	Bad	56	11
## 29	2.99	103	74	0	359	97	Bad	55	11
## 299	10.98	148	63	0	312	130	Good	63	15
## 297	8.21	127	44	13	160	123	Good	63	18
## 124	8.19	127	103	0	125	155	Good	29	15
## 106	5.55	104	100	8	398	97	Medium	61	11
## 231	5.16	115	60	0	119	114	Bad	38	14
## 315	7.72	133	33	10	333	129	Good	71	14
## 389	8.14	135	89	11	245	78	Bad	79	16
## 115	9.31	122	87	9	17	106	Medium	65	13
## 23	5.08	128	46	6	497	138	Medium	42	13
## 387	5.32	152	116	0	170	160	Medium	39	16
## 383	4.95	121	28	19	315	121	Medium	66	14
## 143	7.44	124	84	0	300	104	Medium	77	15
## 317	15.63	122	36	5	369	72	Good	35	10
## 129	4.96	133	100	3	350	126	Bad	55	13
## 57	11.91	133	82	0	54	84	Medium	50	17
## 151	10.49	122	84	8	176	114	Good	57	10
## 170	11.48	104	41	15	492	77	Good	73	18
## 217	5.73	141	33	0	243	144	Medium	34	17
## 179	10.66	104	71	14	89	81	Medium	25	14
## 2	11.22	111	48	16	260	83	Good	65	10
## 203	4.10	121	78	4	413	130	Bad	46	10
## 234	8.65	123	76	18	218	120	Medium	29	14
##	Urban	US							
## 216	Yes	Yes							
## 316	Yes	Yes							
## 103	No	No							
## 20	Yes	Yes							
## 268	No	Yes							
## 167	Yes	Yes							
## 63	Yes	Yes							
## 17	Yes	No							
## 180	Yes	Yes							
## 83	Yes	Yes							
## 264	Yes	Yes							
## 225	No	No							
## 304	Yes	Yes							
## 53	Yes	Yes							
## 118	Yes	No							
## 91	No	No							
## 300	No	Yes							

## 80	Yes	No
## 322	Yes	No
## 172	Yes	Yes
## 330	Yes	Yes
## 195	Yes	Yes
## 43	Yes	No
## 265	Yes	Yes
## 341	Yes	No
## 161	No	No
## 215	Yes	Yes
## 187	No	No
## 223	Yes	Yes
## 122	Yes	Yes
## 396	Yes	Yes
## 248	Yes	No
## 303	Yes	Yes
## 166	Yes	Yes
## 272	Yes	No
## 274	Yes	Yes
## 386	Yes	Yes
## 336	Yes	Yes
## 14	Yes	Yes
## 135	Yes	No
## 132	Yes	No
## 184	Yes	Yes
## 327	Yes	No
## 399	Yes	Yes
## 45	Yes	Yes
## 70	Yes	No
## 360	Yes	Yes
## 22	No	Yes
## 154	No	Yes
## 235	No	Yes
## 69	Yes	Yes
## 397	No	Yes
## 230	No	No
## 159	No	Yes
## 319	No	Yes
## 148	No	Yes
## 153	No	No
## 13	Yes	No
## 352	No	Yes
## 178	Yes	Yes
## 102	Yes	No
## 227	Yes	No
## 362	No	Yes
## 140	No	Yes
## 107	No	No
## 120	Yes	Yes
## 86	No	No
## 30	Yes	Yes
## 33	No	Yes
## 54	Yes	Yes
## 44	Yes	Yes

## 266	Yes	Yes
## 121	Yes	Yes
## 308	Yes	No
## 68	Yes	Yes
## 261	Yes	Yes
## 160	No	No
## 271	Yes	No
## 7	Yes	No
## 366	No	No
## 357	Yes	No
## 245	Yes	No
## 393	Yes	Yes
## 329	Yes	Yes
## 334	Yes	Yes
## 371	Yes	Yes
## 346	Yes	No
## 350	No	Yes
## 198	Yes	No
## 291	No	Yes
## 333	Yes	Yes
## 110	No	No
## 295	Yes	Yes
## 354	No	Yes
## 260	No	Yes
## 192	Yes	Yes
## 128	Yes	Yes
## 36	No	Yes
## 40	No	No
## 116	Yes	No
## 256	Yes	Yes
## 278	Yes	Yes
## 62	No	No
## 27	No	Yes
## 64	Yes	Yes
## 85	No	No
## 326	Yes	Yes
## 61	Yes	Yes
## 37	No	No
## 377	Yes	Yes
## 177	No	Yes
## 165	No	Yes
## 93	Yes	No
## 246	No	Yes
## 254	No	Yes
## 384	Yes	No
## 398	Yes	Yes
## 263	Yes	Yes
## 75	No	Yes
## 220	Yes	Yes
## 318	No	No
## 126	No	No
## 47	No	Yes
## 324	Yes	Yes
## 46	Yes	Yes

## 190	No	Yes
## 19	No	Yes
## 59	Yes	Yes
## 306	Yes	Yes
## 9	No	No
## 214	Yes	Yes
## 292	Yes	No
## 344	Yes	Yes
## 176	Yes	No
## 175	No	No
## 226	Yes	No
## 48	Yes	No
## 222	Yes	No
## 145	No	No
## 355	Yes	Yes
## 185	No	Yes
## 74	No	Yes
## 212	Yes	Yes
## 119	Yes	Yes
## 207	Yes	Yes
## 34	Yes	Yes
## 325	Yes	Yes
## 5	Yes	No
## 197	Yes	Yes
## 186	Yes	Yes
## 193	No	No
## 182	Yes	No
## 65	No	Yes
## 285	No	No
## 284	No	No
## 365	Yes	Yes
## 381	Yes	Yes
## 31	Yes	No
## 117	No	No
## 244	Yes	Yes
## 88	No	Yes
## 349	Yes	Yes
## 395	Yes	Yes
## 309	Yes	Yes
## 363	Yes	Yes
## 183	Yes	No
## 221	Yes	Yes
## 28	Yes	No
## 67	Yes	No
## 262	Yes	Yes
## 289	No	No
## 356	Yes	No
## 267	No	Yes
## 378	No	No
## 52	Yes	No
## 150	Yes	Yes
## 279	No	Yes
## 282	No	Yes
## 4	Yes	Yes

## 276	Yes	Yes
## 376	Yes	No
## 209	Yes	No
## 373	No	No
## 173	Yes	Yes
## 269	Yes	No
## 290	Yes	Yes
## 127	Yes	Yes
## 155	No	Yes
## 109	Yes	No
## 380	Yes	No
## 136	No	Yes
## 49	Yes	No
## 194	Yes	Yes
## 255	Yes	Yes
## 125	Yes	No
## 302	Yes	Yes
## 157	Yes	No
## 233	Yes	Yes
## 8	Yes	Yes
## 239	Yes	No
## 156	Yes	No
## 123	Yes	Yes
## 249	Yes	Yes
## 331	No	No
## 108	Yes	No
## 208	No	No
## 96	Yes	Yes
## 385	Yes	Yes
## 144	Yes	Yes
## 335	Yes	Yes
## 164	No	No
## 205	Yes	No
## 210	No	Yes
## 112	Yes	Yes
## 258	Yes	Yes
## 16	No	No
## 394	No	Yes
## 113	Yes	Yes
## 97	No	Yes
## 277	Yes	Yes
## 364	Yes	No
## 275	Yes	Yes
## 196	Yes	Yes
## 296	No	Yes
## 270	Yes	No
## 342	No	No
## 340	Yes	Yes
## 41	No	No
## 305	Yes	Yes
## 242	Yes	No
## 391	Yes	Yes
## 72	No	Yes
## 90	No	No

## 358	Yes	Yes
## 219	Yes	Yes
## 312	Yes	Yes
## 58	Yes	No
## 73	Yes	No
## 152	No	Yes
## 206	Yes	No
## 139	Yes	Yes
## 92	Yes	Yes
## 163	Yes	No
## 369	No	Yes
## 89	Yes	Yes
## 55	No	Yes
## 241	Yes	No
## 237	Yes	Yes
## 345	No	Yes
## 367	No	Yes
## 94	Yes	No
## 82	Yes	No
## 370	Yes	Yes
## 98	Yes	Yes
## 56	Yes	Yes
## 24	Yes	No
## 228	Yes	Yes
## 343	No	Yes
## 15	Yes	Yes
## 181	Yes	Yes
## 359	Yes	Yes
## 347	No	No
## 104	Yes	Yes
## 146	Yes	Yes
## 42	Yes	No
## 162	No	Yes
## 390	Yes	Yes
## 3	Yes	Yes
## 38	Yes	Yes
## 131	Yes	Yes
## 280	Yes	Yes
## 321	Yes	Yes
## 10	No	Yes
## 286	Yes	Yes
## 320	No	Yes
## 95	Yes	Yes
## 323	Yes	Yes
## 240	Yes	Yes
## 361	No	Yes
## 310	Yes	Yes
## 158	No	Yes
## 142	Yes	No
## 200	Yes	Yes
## 204	Yes	No
## 273	Yes	No
## 259	No	No
## 287	No	Yes



## 332	Yes	Yes
## 100	No	Yes
## 11	No	Yes
## 50	Yes	No
## 81	Yes	Yes
## 77	Yes	Yes
## 281	Yes	Yes
## 169	Yes	No
## 76	No	Yes
## 372	Yes	No
## 99	No	Yes
## 134	Yes	Yes
## 388	No	Yes
## 141	Yes	Yes
## 1	Yes	Yes
## 66	No	No
## 301	Yes	Yes
## 339	Yes	No
## 211	No	Yes
## 149	No	Yes
## 137	No	No
## 252	Yes	Yes
## 253	Yes	No
## 375	Yes	Yes
## 71	Yes	Yes
## 79	Yes	Yes
## 130	No	Yes
## 400	Yes	Yes
## 293	Yes	Yes
## 12	Yes	Yes
## 224	Yes	Yes
## 218	No	No
## 243	No	No
## 171	Yes	Yes
## 251	Yes	Yes
## 247	Yes	Yes
## 111	Yes	Yes
## 147	Yes	No
## 26	No	No
## 288	Yes	Yes
## 168	Yes	No
## 133	Yes	Yes
## 32	Yes	Yes
## 25	Yes	Yes
## 51	Yes	Yes
## 368	Yes	No
## 189	Yes	No
## 202	Yes	No
## 188	Yes	No
## 294	Yes	No
## 199	Yes	Yes
## 39	Yes	No
## 374	Yes	No
## 257	Yes	No

## 60	Yes	No
## 392	Yes	No
## 307	Yes	Yes
## 298	Yes	Yes
## 250	Yes	No
## 6	No	Yes
## 78	No	Yes
## 283	Yes	No
## 238	Yes	Yes
## 337	Yes	No
## 101	No	Yes
## 353	Yes	Yes
## 382	Yes	Yes
## 138	Yes	No
## 21	Yes	Yes
## 351	No	Yes
## 313	Yes	Yes
## 236	Yes	Yes
## 328	Yes	Yes
## 232	No	No
## 348	No	No
## 338	Yes	No
## 105	Yes	No
## 87	Yes	No
## 114	Yes	Yes
## 311	Yes	Yes
## 201	No	No
## 229	No	Yes
## 174	Yes	Yes
## 379	Yes	Yes
## 18	Yes	Yes
## 213	Yes	Yes
## 191	No	Yes
## 35	Yes	Yes
## 314	Yes	No
## 84	Yes	Yes
## 29	Yes	Yes
## 299	Yes	No
## 297	Yes	Yes
## 124	No	Yes
## 106	Yes	Yes
## 231	No	No
## 315	Yes	Yes
## 389	Yes	Yes
## 115	Yes	Yes
## 23	Yes	No
## 387	Yes	No
## 383	Yes	Yes
## 143	Yes	No
## 317	Yes	Yes
## 129	Yes	Yes
## 57	Yes	No
## 151	No	Yes
## 170	Yes	Yes

```
## 217   Yes  No
## 179    No Yes
## 2     Yes Yes
## 203    No Yes
## 234    No Yes
```

```
summary(Carseats)
```

```
##      Sales      CompPrice      Income      Advertising
##  Min.   : 0.000   Min.   : 77   Min.   : 21.00   Min.   : 0.000
## 1st Qu.: 5.390   1st Qu.:115   1st Qu.: 42.75   1st Qu.: 0.000
##  Median : 7.490   Median :125   Median : 69.00   Median : 5.000
##  Mean   : 7.496   Mean   :125   Mean   : 68.66   Mean   : 6.635
## 3rd Qu.: 9.320   3rd Qu.:135   3rd Qu.: 91.00   3rd Qu.:12.000
##  Max.   :16.270   Max.   :175   Max.   :120.00   Max.   :29.000
##  Population      Price      ShelveLoc      Age      Education
##  Min.   : 10.0   Min.   : 24.0   Bad   : 96   Min.   :25.00   Min.   :10.0
## 1st Qu.:139.0   1st Qu.:100.0   Good  : 85   1st Qu.:39.75   1st Qu.:12.0
##  Median :272.0   Median :117.0   Medium:219   Median :54.50   Median :14.0
##  Mean   :264.8   Mean   :115.8                      Mean   :53.32   Mean   :13.9
## 3rd Qu.:398.5   3rd Qu.:131.0                      3rd Qu.:66.00   3rd Qu.:16.0
##  Max.   :509.0   Max.   :191.0                      Max.   :80.00   Max.   :18.0
## Urban      US
## No :118    No :142
## Yes:282   Yes:258
##
##
##
##
```

```
head(Carseats)
```

```
##      Sales CompPrice Income Advertising Population Price ShelveLoc Age Education
## 216  2.34      116     83         15         170   144      Bad   71         11
## 316  6.39      131     21         8         220   171      Good  29         14
## 103  5.30      113     22         0          57    97      Medium 65         16
## 20   8.73      129     76        16          58   121      Medium 69         12
## 268  5.83      134     82         7         473   112      Bad   51         12
## 167  6.71      119     67        17         151   137      Medium 55         11
##      Urban  US
## 216   Yes  Yes
## 316   Yes  Yes
## 103    No  No
## 20    Yes  Yes
## 268    No  Yes
## 167   Yes  Yes
```

```
split <- sample.split(Carseats, SplitRatio = 0.8)
train_cl <- subset(Carseats, split == "TRUE")
test_cl <- subset(Carseats, split == "FALSE")
train_scale <- scale(train_cl[, 1:4])
test_scale <- scale(test_cl[, 1:4])
```

```

classifier_knn <- knn(train = train_scale,
                      test = test_scale,
                      cl = train_cl$Urban,
                      k = 1)
cm <- table(test_cl$Urban, classifier_knn)
cm

```

```

##      classifier_knn
##      No Yes
## No   11  18
## Yes  22  58

```

```

misClassError <- mean(classifier_knn != test_cl$Urban)
print(paste('Accuracy =', 1-misClassError))

```

```

## [1] "Accuracy = 0.63302752293578"

```

```

accuracies <- vector()
errors<-vector()
for(i in 1:25){
  print(paste("For k = ",i))
  classifier_knn <- knn(train = train_scale,
                        test = test_scale,
                        cl = train_cl$Urban,
                        k = i)

  misClassError <- mean(classifier_knn != test_cl$Urban)
  print(paste('Accuracy =', 1-misClassError))
  print(paste('error =', misClassError))
  accuracies[i] <- 1-misClassError
  errors[i]<-misClassError
}

```

```

## [1] "For k = 1"
## [1] "Accuracy = 0.63302752293578"
## [1] "error = 0.36697247706422"
## [1] "For k = 2"
## [1] "Accuracy = 0.605504587155963"
## [1] "error = 0.394495412844037"
## [1] "For k = 3"
## [1] "Accuracy = 0.697247706422018"
## [1] "error = 0.302752293577982"
## [1] "For k = 4"
## [1] "Accuracy = 0.669724770642202"
## [1] "error = 0.330275229357798"
## [1] "For k = 5"
## [1] "Accuracy = 0.678899082568807"
## [1] "error = 0.321100917431193"
## [1] "For k = 6"
## [1] "Accuracy = 0.715596330275229"
## [1] "error = 0.284403669724771"
## [1] "For k = 7"
## [1] "Accuracy = 0.724770642201835"

```

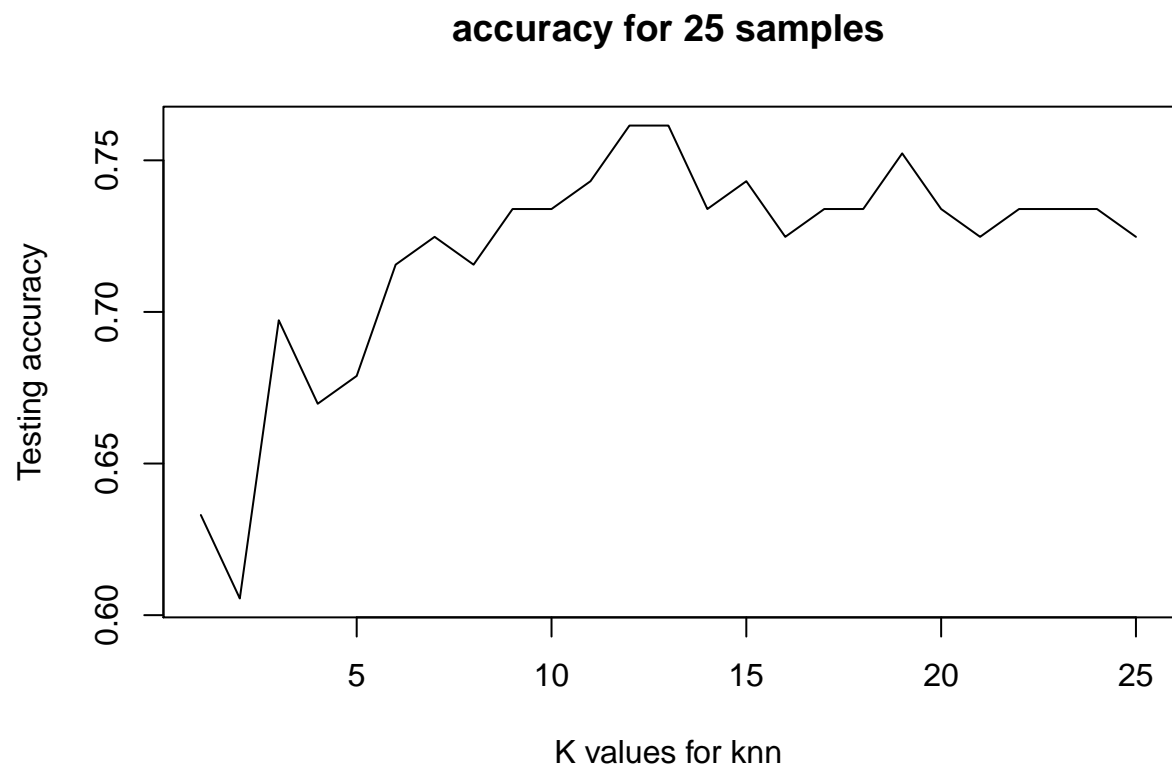
```

## [1] "error = 0.275229357798165"
## [1] "For k = 8"
## [1] "Accuracy = 0.715596330275229"
## [1] "error = 0.284403669724771"
## [1] "For k = 9"
## [1] "Accuracy = 0.73394495412844"
## [1] "error = 0.26605504587156"
## [1] "For k = 10"
## [1] "Accuracy = 0.73394495412844"
## [1] "error = 0.26605504587156"
## [1] "For k = 11"
## [1] "Accuracy = 0.743119266055046"
## [1] "error = 0.256880733944954"
## [1] "For k = 12"
## [1] "Accuracy = 0.761467889908257"
## [1] "error = 0.238532110091743"
## [1] "For k = 13"
## [1] "Accuracy = 0.761467889908257"
## [1] "error = 0.238532110091743"
## [1] "For k = 14"
## [1] "Accuracy = 0.73394495412844"
## [1] "error = 0.26605504587156"
## [1] "For k = 15"
## [1] "Accuracy = 0.743119266055046"
## [1] "error = 0.256880733944954"
## [1] "For k = 16"
## [1] "Accuracy = 0.724770642201835"
## [1] "error = 0.275229357798165"
## [1] "For k = 17"
## [1] "Accuracy = 0.73394495412844"
## [1] "error = 0.26605504587156"
## [1] "For k = 18"
## [1] "Accuracy = 0.73394495412844"
## [1] "error = 0.26605504587156"
## [1] "For k = 19"
## [1] "Accuracy = 0.752293577981651"
## [1] "error = 0.247706422018349"
## [1] "For k = 20"
## [1] "Accuracy = 0.73394495412844"
## [1] "error = 0.26605504587156"
## [1] "For k = 21"
## [1] "Accuracy = 0.724770642201835"
## [1] "error = 0.275229357798165"
## [1] "For k = 22"
## [1] "Accuracy = 0.73394495412844"
## [1] "error = 0.26605504587156"
## [1] "For k = 23"
## [1] "Accuracy = 0.73394495412844"
## [1] "error = 0.26605504587156"
## [1] "For k = 24"
## [1] "Accuracy = 0.73394495412844"
## [1] "error = 0.26605504587156"
## [1] "For k = 25"
## [1] "Accuracy = 0.724770642201835"

```

```
## [1] "error = 0.275229357798165"
```

```
plot(1:25,accuracies,type="l",ylab=" Testing accuracy ",xlab = "K values for knn",main = "accuracy for 25 samples")
```



```
plot(1:25,errors,type="l",ylab=" error rate ",xlab = "K values for knn",main = "error for 25 samples")
```

